

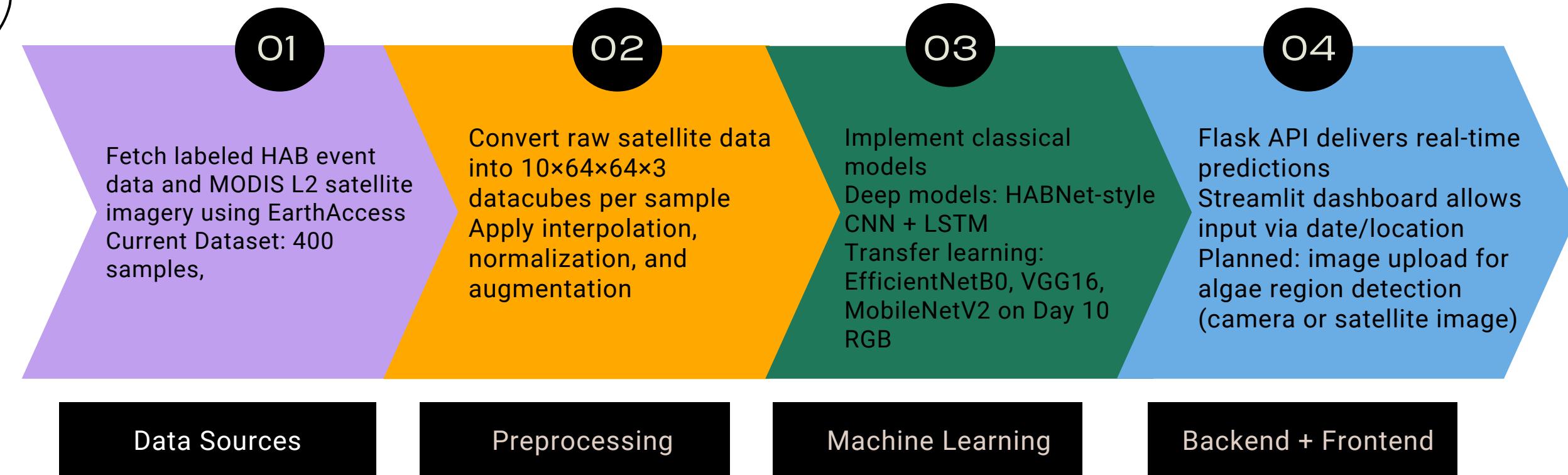
# Harmful Algae Bloom Detection System

Team: Gradient Descents

# Objectives

- Develop a machine learning pipeline to classify HAB events as Toxic or Non-Toxic using multi-day, multi-modality satellite imagery (10-day sequences, 3 modalities)
- Train and evaluate diverse models, including classical ML , deep learning, and transfer learning
- Analyze temporal-spatial patterns and use augmentation to enhance feature learning and interpretability
- Begin integrating real-time capabilities, including prediction interface development and groundwork for algae region detection from uploaded imagery

# System Architecture



## Additions:

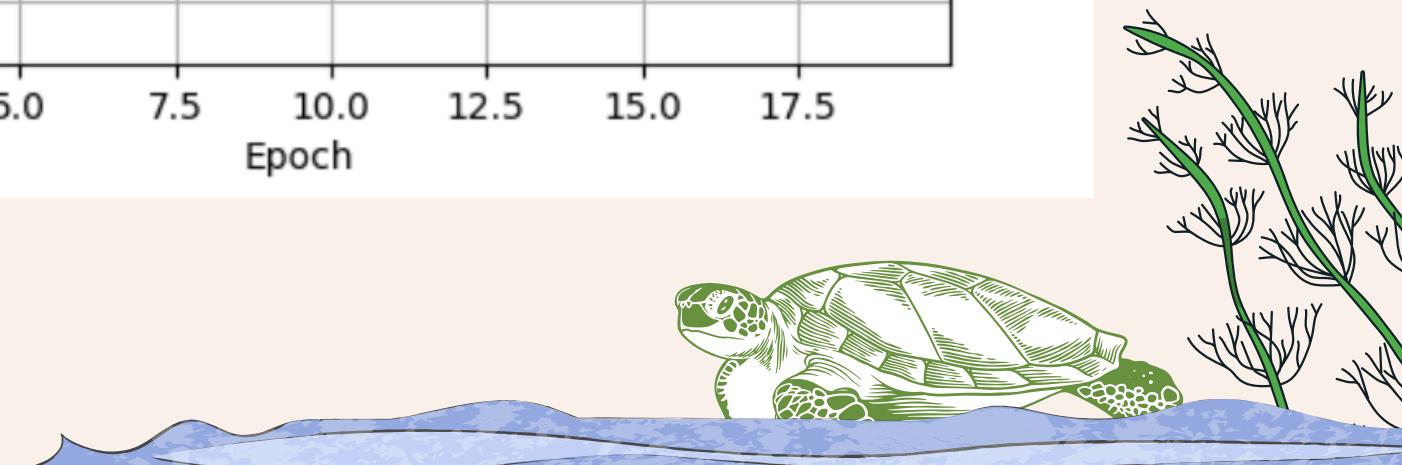
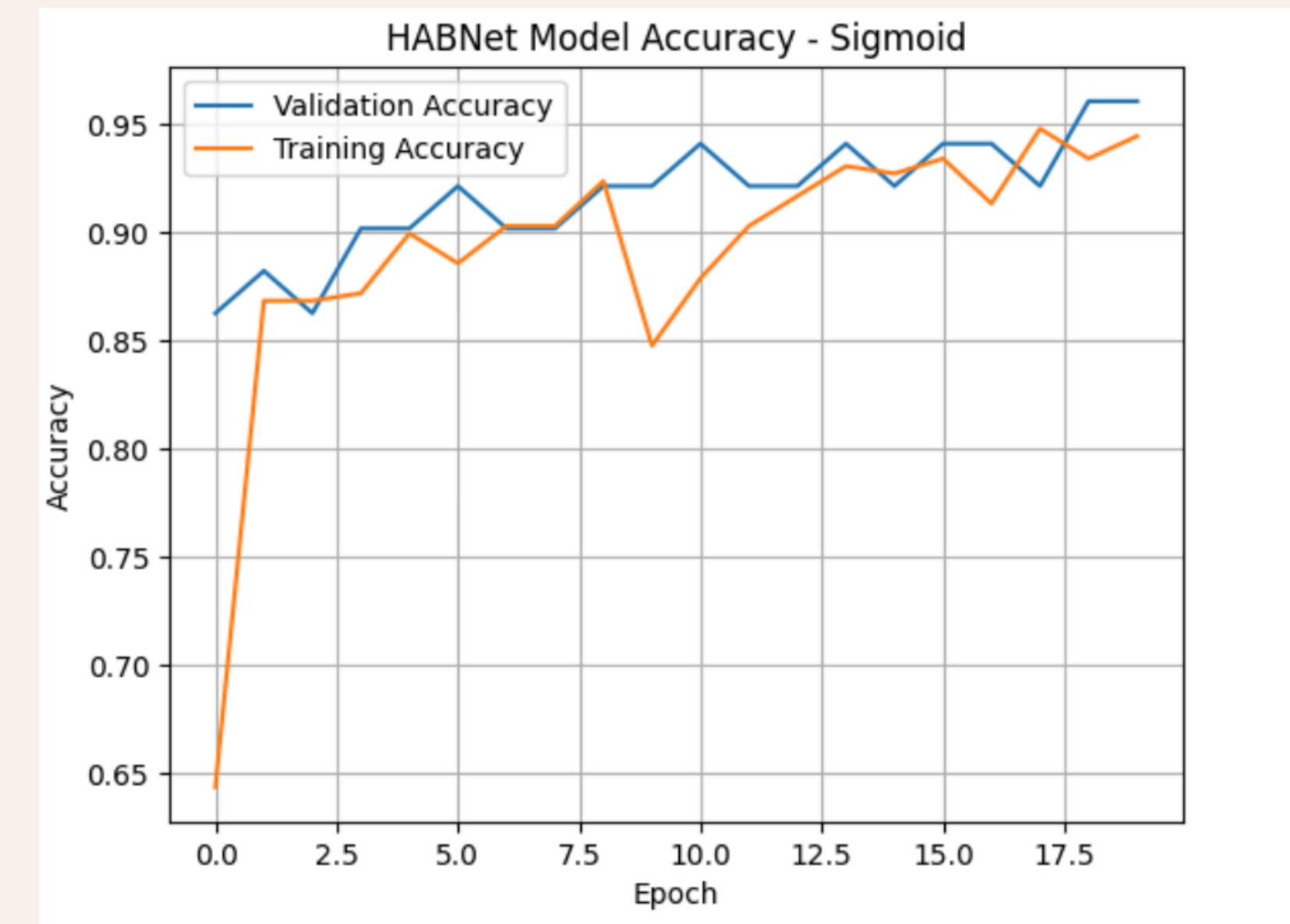
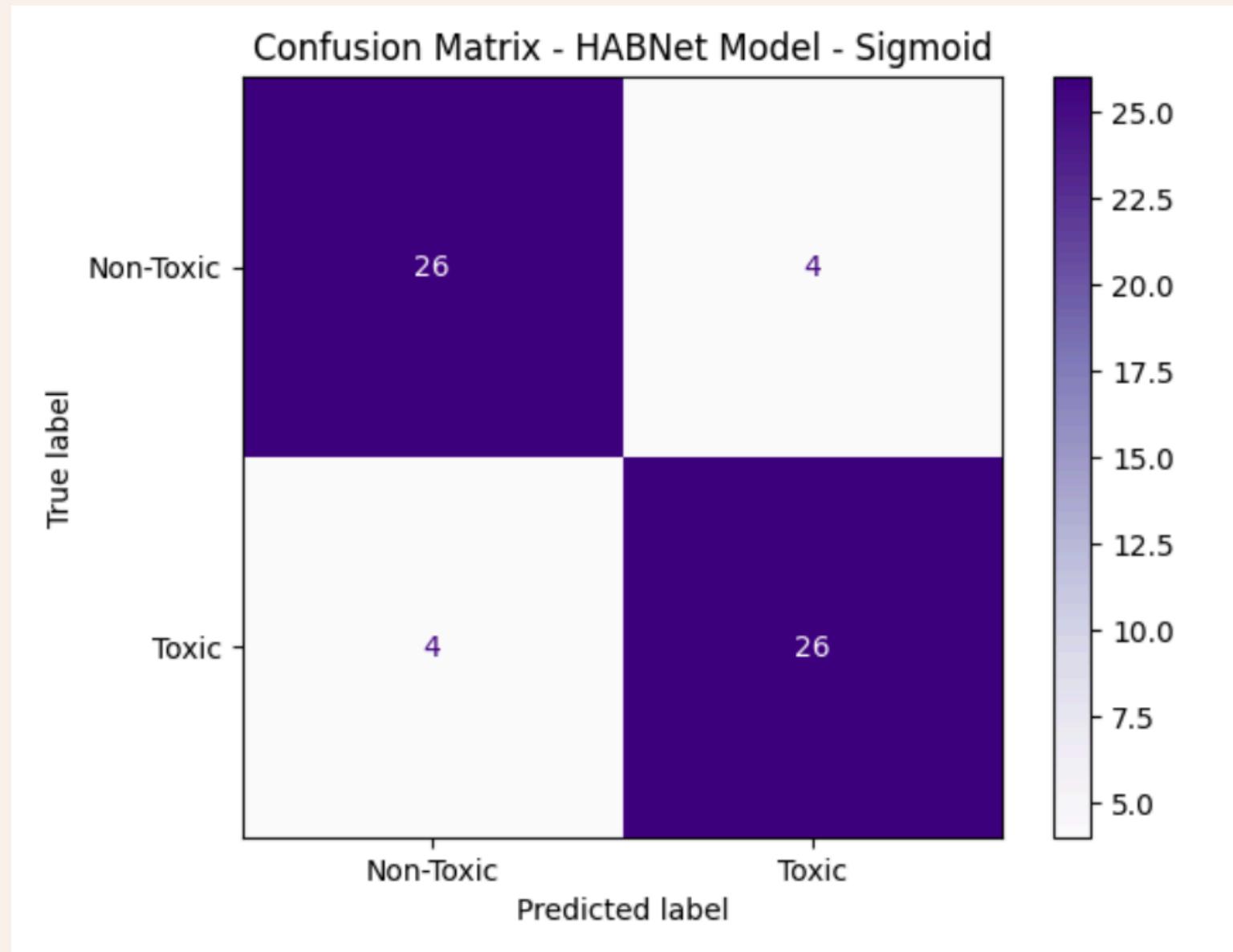
Interactive dashboard connects with Flask API and supports both event-based and manual inputs

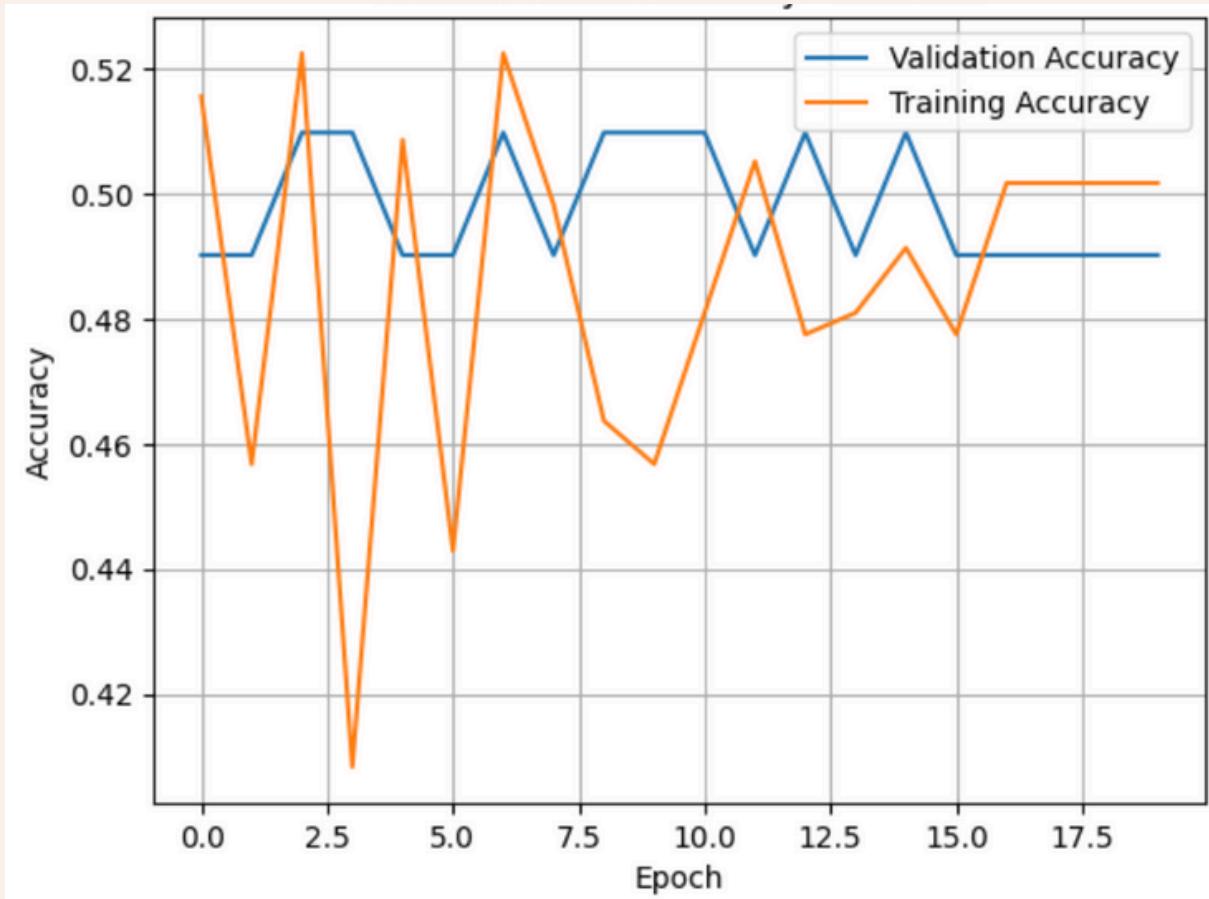
# Progress Since Last Presentation

- Developed a full ML pipeline to classify HAB events as Toxic or Non-Toxic using 10-day, 3-modality satellite imagery
- Trained multiple models: baseline ML (Logistic Regression, Ridge, XGBoost), deep learning (CNN + LSTM), and transfer learning (EfficientNetB0, VGG16, MobileNetV2)
- Demonstrated solid accuracy with the HABNet-style CNN + LSTM model on limited data
- Added augmentation to improve robustness and understand feature separability

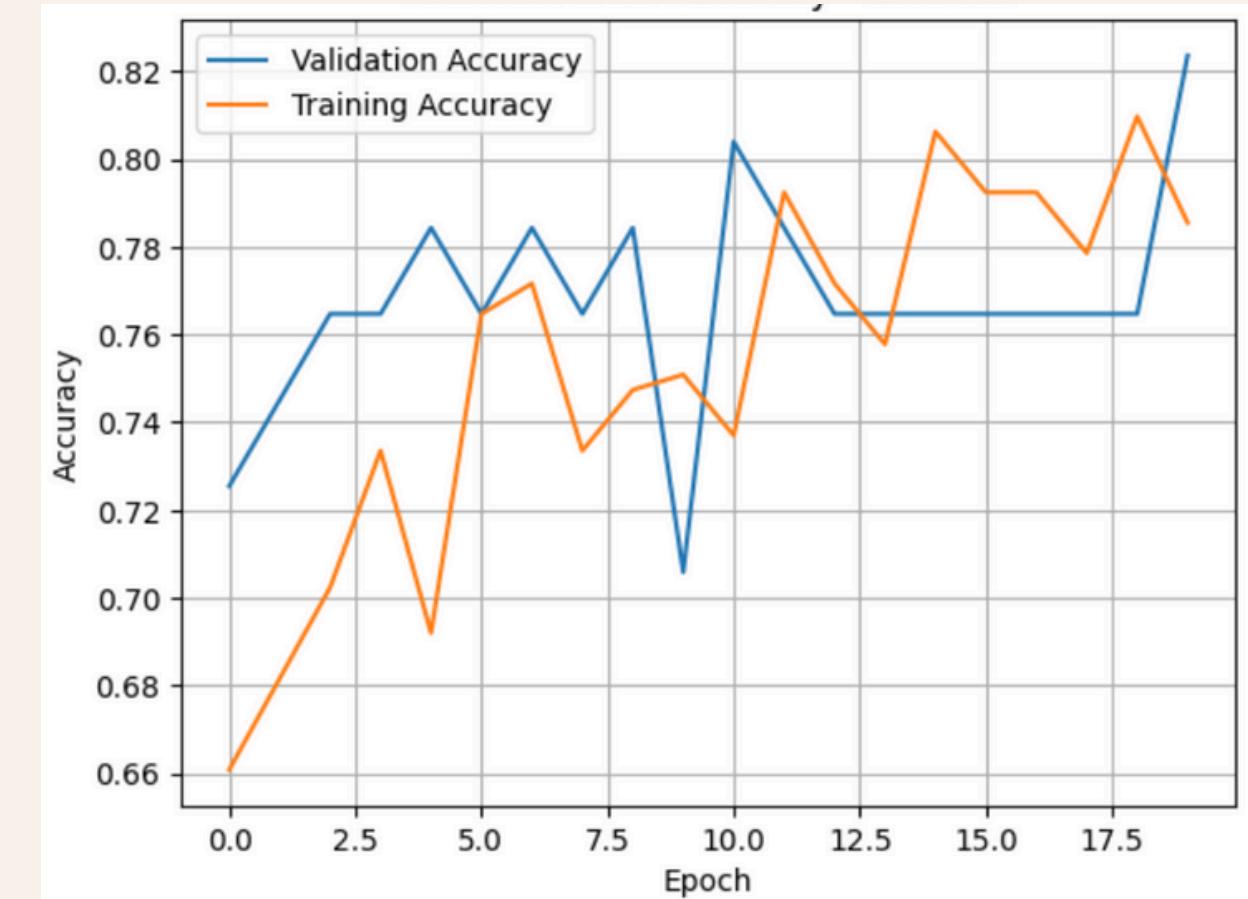


# Machine Learning

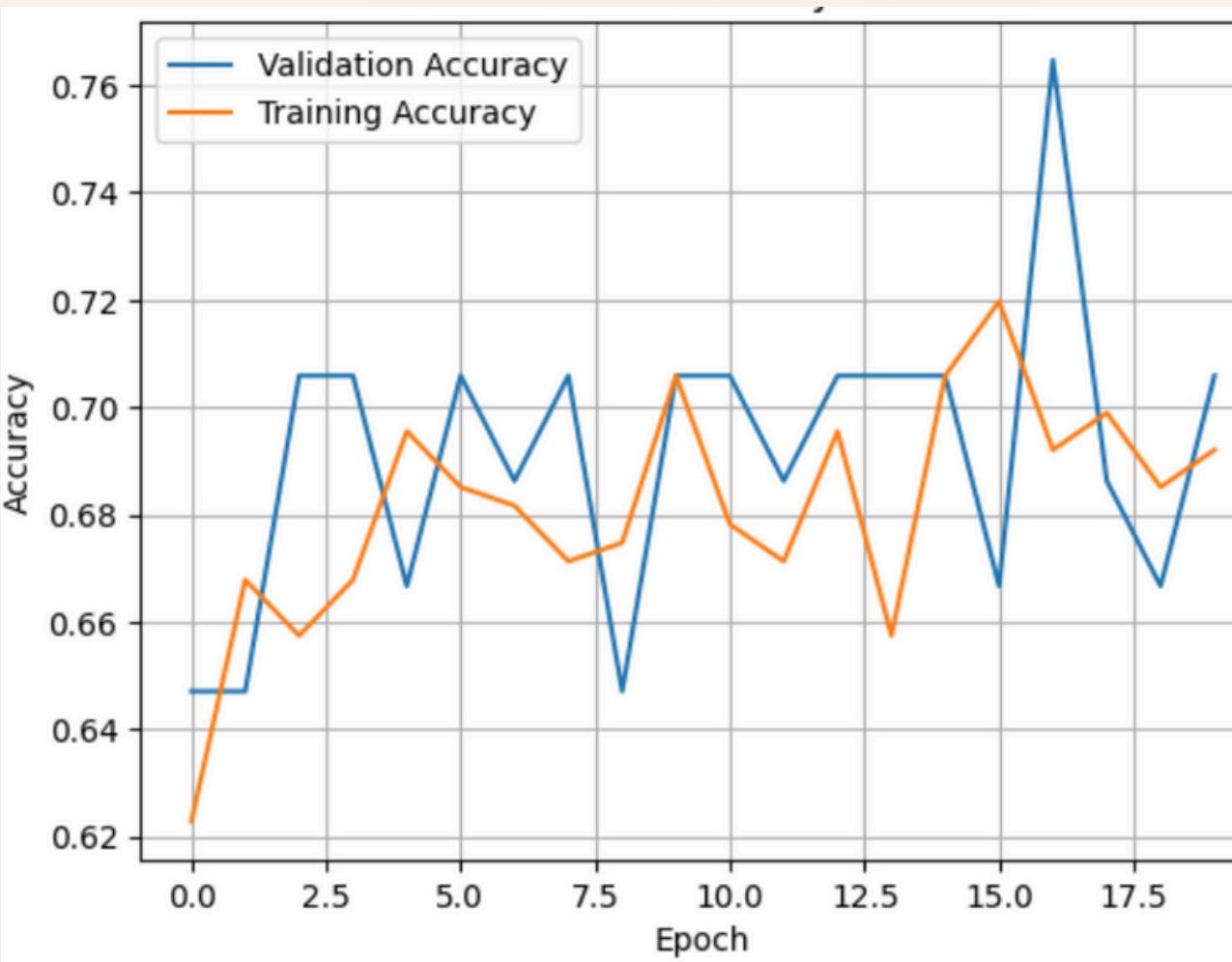




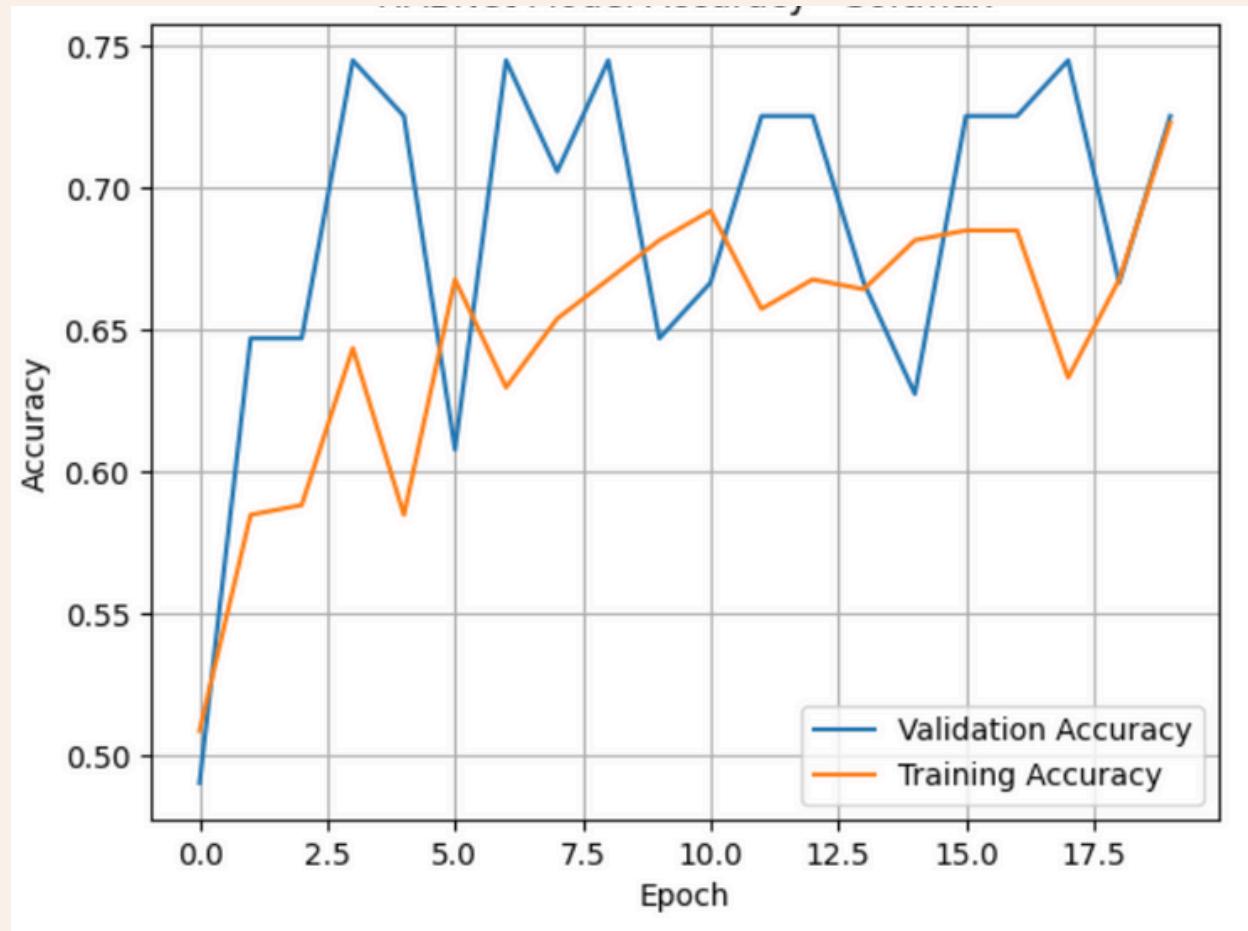
**EfficientNetB0**



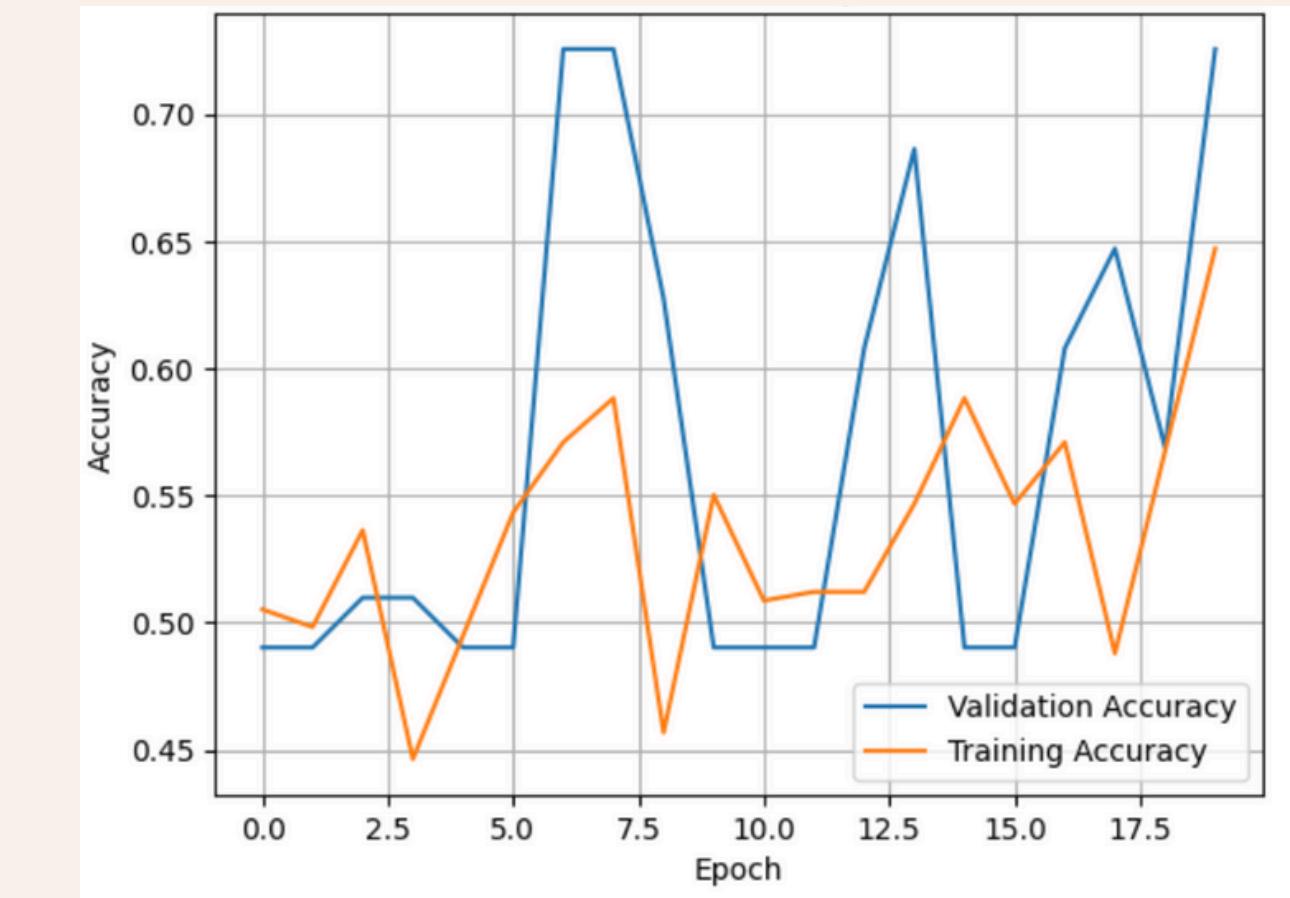
**MobileNetV2**



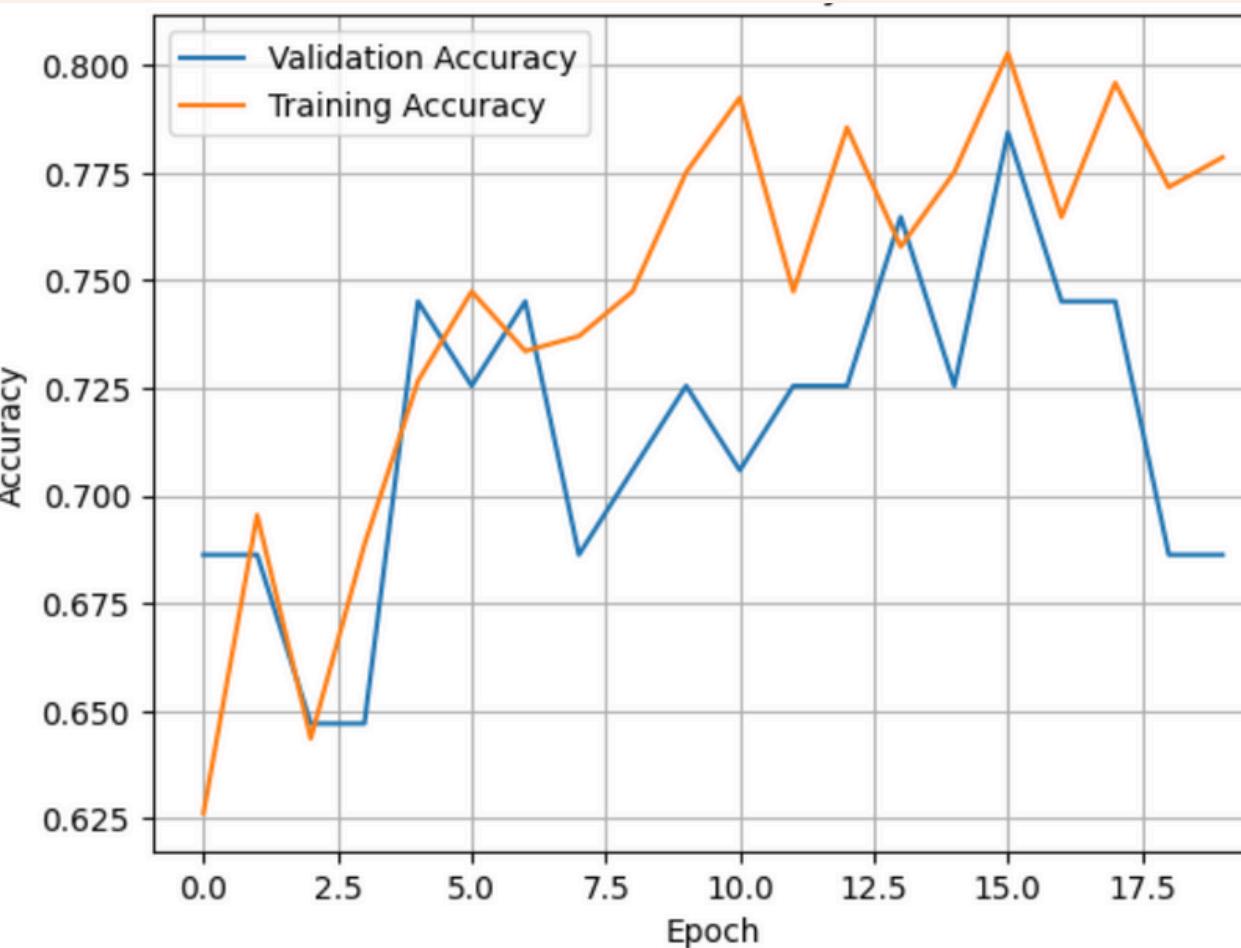
**VGG16**



VGG19



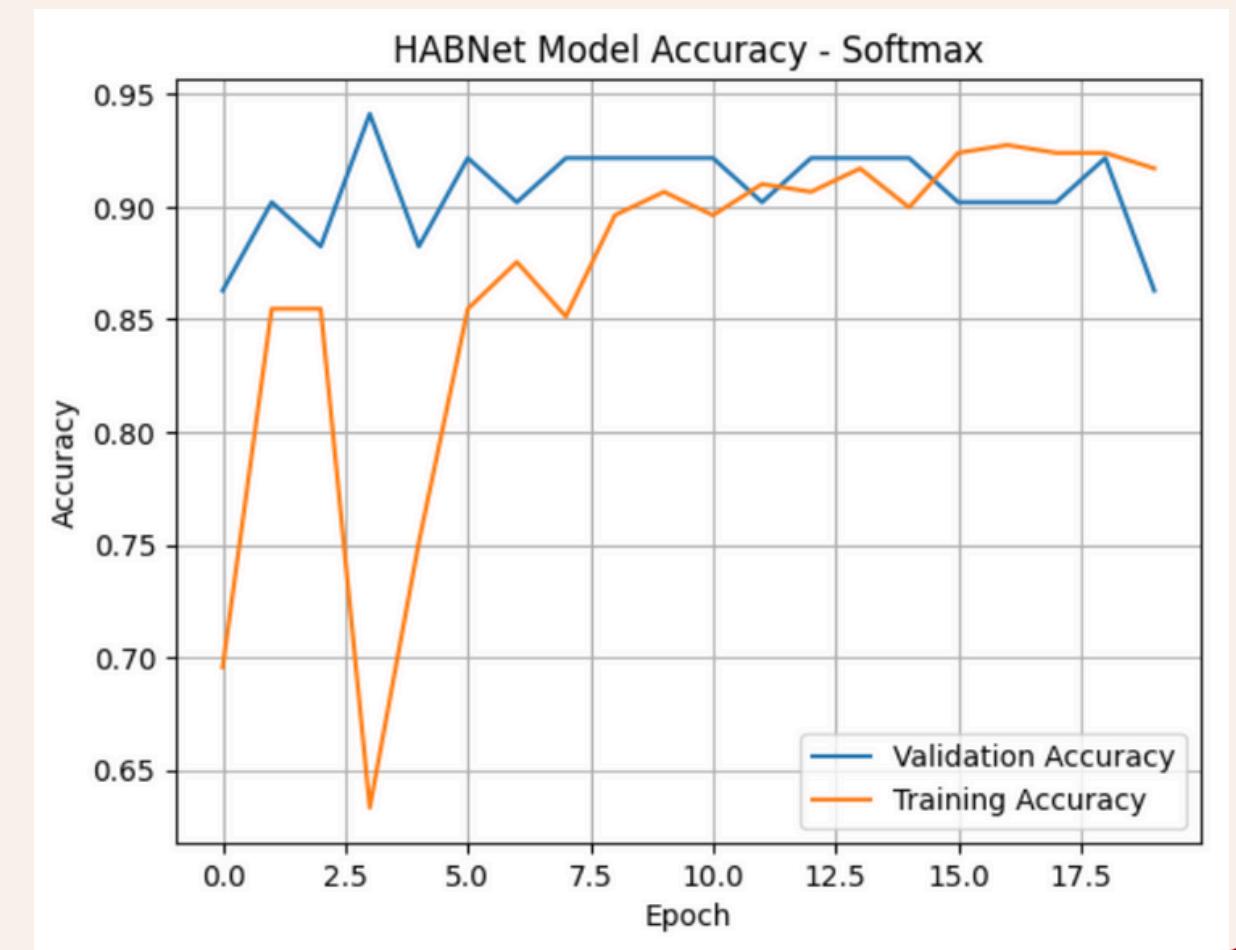
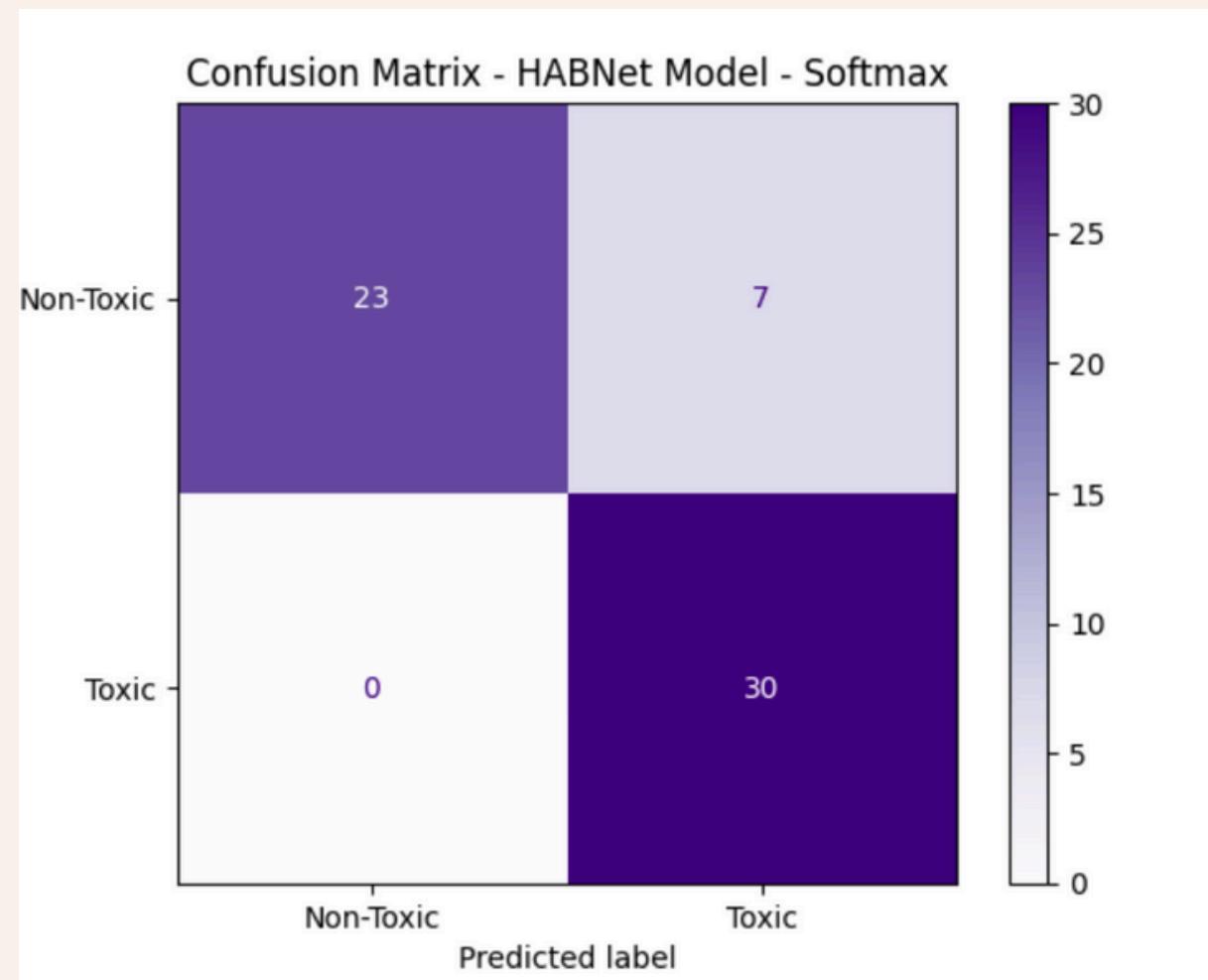
ResNet50



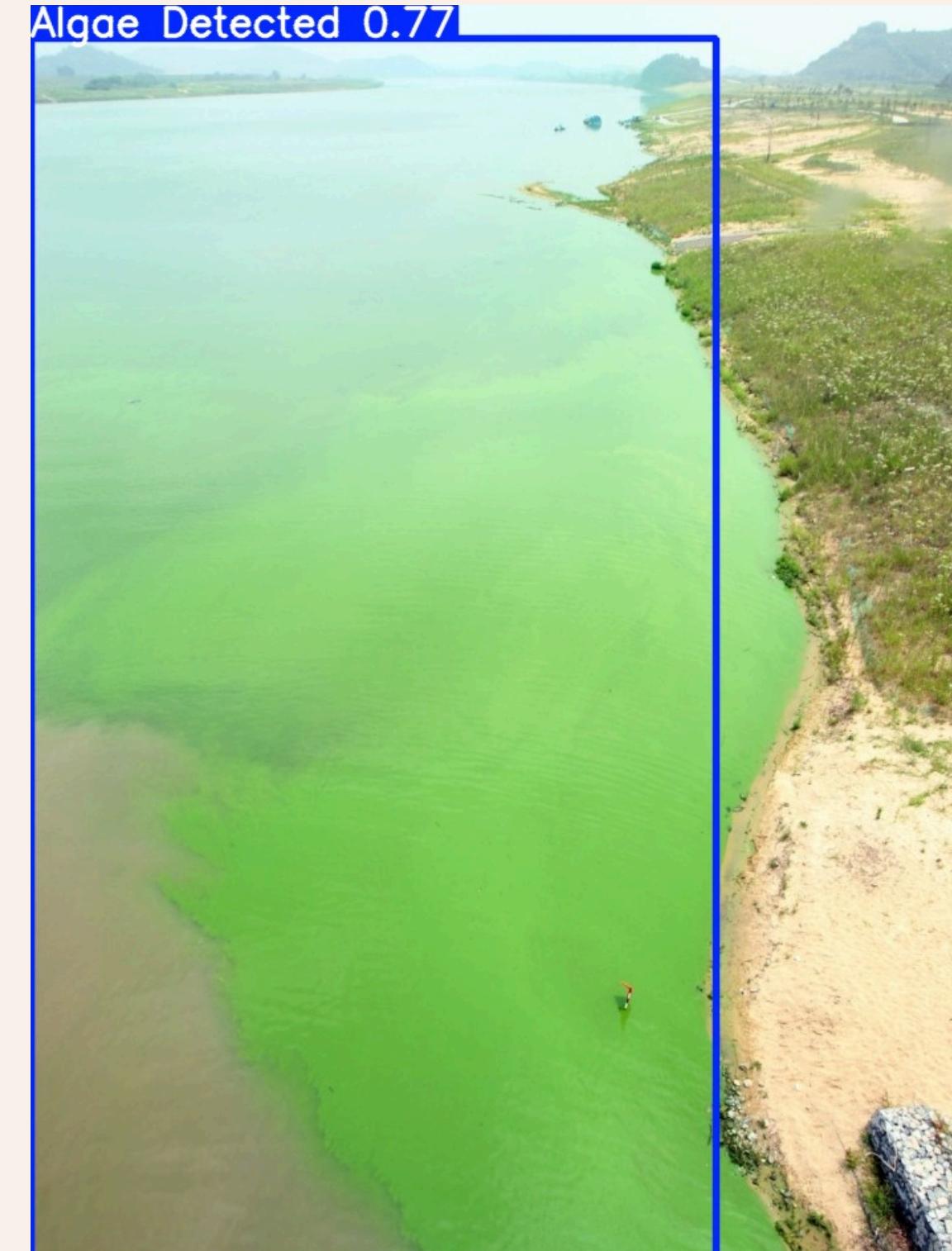
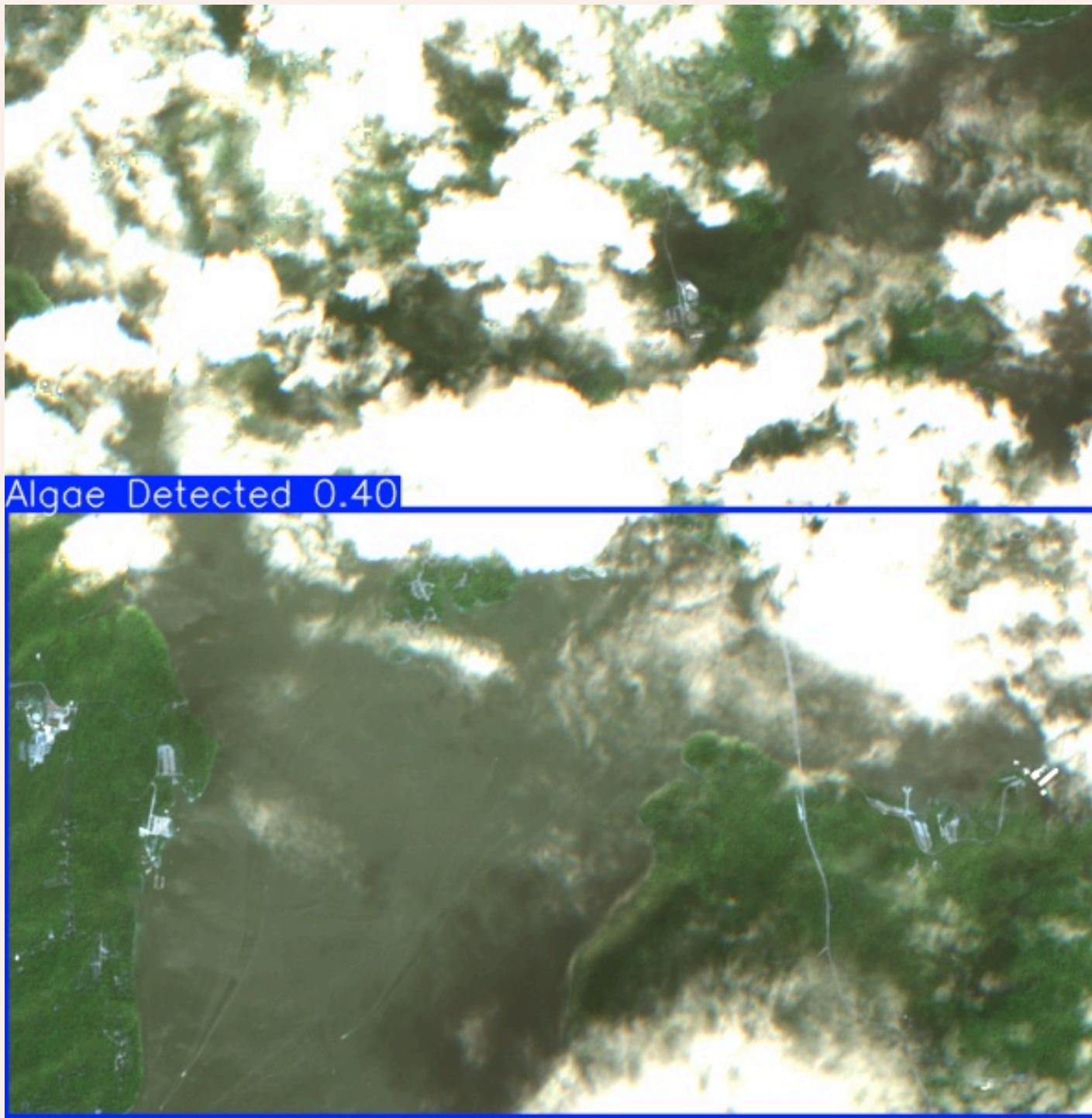
NASNetMobile

# Machine Learning

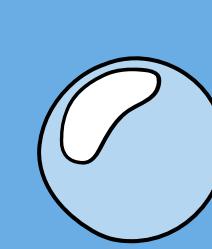
	Model	Accuracy
0	Logistic Regression	0.92
1	Ridge Classifier	0.90
2	XGB Classifier	0.85
3	DT Classifier	0.85
4	RF Classifier	0.88
5	HABnet Sigmoid	0.87
6	HABnet Softmax	0.88
7	EfficientNet	0.50
8	MobileNet	0.77
9	VGGG16	0.75
10	VGG19	0.73
11	ResNet	0.73
12	NasNet	0.70



# Added Interactivities



# Plans for Coming Week



## Pipeline, Backend & Frontend

- Optimize the end-to-end workflow to reduce overhead in data extraction and prediction latency
- Deploy the full pipeline on a cloud platform, preferably AWS, for scalable access
- Begin adding image upload functionality to the frontend for detecting algae regions from both camera and satellite images

## ML Experiments

- Focus on reducing false negatives by improving recall in toxic bloom classification
- Experiment with pre-trained CNNs as feature extractors and combine them with a custom classification neural network
- Compare performance against existing models (e.g., HABNet, XGBoost) for generalization and robustness

# Overall Accomplishments

- Developed an end-to-end ML pipeline for classifying HAB events as Toxic or Non-Toxic using 10-day, 3-modality satellite datacubes
- Implemented and evaluated multiple models: baseline classifiers (Logistic Regression, Ridge, XGBoost), deep learning (HABNet-style CNN + LSTM), and transfer learning (EfficientNetB0, VGG16, MobileNetV2)
- Demonstrated strong performance from CNN + LSTM on limited data and high accuracy (>85%) using Day-10 RGB with pretrained CNNs
- Used augmentation for improved model robustness and better class separability
- Laid groundwork for real-time prediction interface, enabling future user interaction and field usage

# Review and Tasks Assigned

## CHALLENGE

TASKS	ASSIGNEE
Data collection and preprocessing	Daniel
Machine learning and Model Training	Kruthi, Dharmik
Backend	Sagar, Karthika
Frontend	Roshan

# Thank You

